

A VECTOR ESTIMATION SYSTEM, METHOD AND ASSOCIATED ENCODER

FIELD OF THE INVENTION

5 This invention relates to an encoder and a vector estimation system and method for processing a sequence of vectors to determine a filtered estimate value of each vector. The invention is particularly useful for, but not necessarily limited to, determining filtered estimate
10 values to be encoded by a speech encoder and transmitted over a communication link.

BACKGROUND ART

15 A digital speech communication or storage system typically uses a speech encoder to produce a parsimonious representation of the speech signal. A corresponding decoder is used to generate an approximation to the speech signal from that representation. The combination of the
20 encoder and decoder is known in the art as a speech codec. As will be apparent to a person skilled in the art, many segments of speech signals contain quasiperiodic waveforms. Accordingly, consecutive cycles of quasiperiodic waveforms can be considered, and processed, by a speech codec as data
25 vectors that evolve slowly over time.

 An important element of a speech codec is the way it exploits correlation between consecutive cycles of quasiperiodic waveforms. Frequently, correlation is
30 exploited by transmitting a single cycle of the waveform, or of a filtered version of the waveform, only once every 20-30 ms, so that a portion of the data is missing in the received signal. In a typical decoder the missing data is

determined by interpolating between samples of the transmitted cycles.

In general, the use of interpolation by a speech
 5 decoder to generate data between the transmitted cycles
 only produces an adequate approximation to the speech
 signal if the speech signal really is quasiperiodic, or,
 equivalently, if the vectors representing consecutive
 cycles of the waveform evolve sufficiently slowly.
 10 However, many segments of speech contain noisy signal
 components, and this results in comparatively rapid
 evolution of the waveform cycles. In order for waveform
 interpolation in an encoder to be useful for such signals,
 it is necessary to extract a sufficiently quasiperiodic
 15 component from the noisy signal in the encoder. This
 extracted component may be encoded by transmitting only
 selected cycles and decoded by interpolation in the manner
 described above. The remaining noisy component may also be
 encoded using other appropriate techniques and combined
 20 with the quasiperiodic component in the decoder.

Linear low pass filtering a sequence of vectors
 representing consecutive cycles of speech in the time
 dimension is well known in the speech coding literature.
 25 The difficulty with this approach is that in order to get
 good separation of the slowly and rapidly evolving
 components, the low pass filter frequency response must
 have a sharp roll-off. This requires a long impulse
 response, which necessitates an undesirably large filter
 30 delay.

A Kalman filter technique for estimating quasiperiodic
 signal components has been described by Gruber and Todtli
 (IEEE Trans Signal Processing, Vol. 42, No. 3, March 1994,

pp 552-562). However, because this Kalman filter technique is based on a linear dynamic system model of a frequency domain representation of the signal, it is unnecessarily complex. It also assumes that the dynamic system model parameters (i.e. noise energy and the harmonic signal gain) are known. However, when considering speech coding, noise energy and the harmonic signal gain parameters are not known.

A technique for determining the system parameters required in a Kalman filter using an Expectation Maximisation algorithm has been described in a more general setting by Digalakis et al (IEEE Trans Speech and Audio Processing, Vol. 1, No. 4, Oct. 1993, pp 431-442). However, the technique is iterative, and in the absence of good initial estimates may converge slowly. It may also produce a result that is not globally optimal. No prior art method is known for obtaining good initial estimates. Further, this method typically requires a significant amount of data, over which the unknown parameters are constant. In the context of speech coding, where the parameters change continuously, rapid estimation is essential, and therefore this method of applying the Expectation Maximization algorithm needs to be improved.

Stachurski (PhD Thesis, McGill University, Montreal Canada, 1997) proposed a technique for estimating quasiperiodic signal components of a speech signal. This method involves minimizing a weighted combination of estimated noise energy and a measure of rate of change in the quasiperiodic component. This method is highly complex and does not allow the rate of evolution of the quasiperiodic component to be specified independently. Nor does it allow for an independently varying gain for the quasiperiodic component.

In this specification, including the claims, the terms comprises, comprising or similar terms are intended to mean a non-exclusive inclusion, such that a method or apparatus
5 that comprises a list of elements does not include those elements solely, but may well include other elements not listed.

SUMMARY OF THE INVENTION

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According to one aspect of the invention there is provided a vector estimation system for processing a sequence of input vectors, said input vectors each comprising a plurality of elements, and said system
15 comprising:

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a digital filter with a filter vector input for receiving said sequence of input vectors and a predictor gain input for controlling characteristics of said filter, said digital filter also having both a
20 current slowly evolving filter estimate output and a previous slowly evolving filter estimate output, said current slowly evolving filter estimate output providing a current filtered estimate value of a slowly evolving component of said sequence of input
25 vectors and said previous slowly evolving filter estimate output providing a previous filtered estimate value of said slowly evolving component of said sequence of input vectors; and

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a parameter estimator having an estimator vector input for receiving said sequence of input vectors and a previous slowly evolving filter estimate input coupled to said previous slowly evolving filter estimate output, said parameter estimator further

includes a predictor gain output coupled to said predictor gain input,

wherein when said vector estimation system receives a current input vector that is one of said sequence of said input vectors, said parameter estimator provides a current predictor gain value at said predictor gain output thereby modifying said current filtered estimate value at said current slowly evolving filter estimate output, said current predictor gain value being dependent upon both said previous filtered estimate value received at said slowly evolving filter estimate input and said current input vector received at said estimator vector input.

Suitably, said parameter estimator is characterised by said current predictor gain value being dependent upon both a sequence of previous said input vectors and a sequence of filtered estimate values provided by successive preceding values of said previous filtered estimate value.

Preferably, said current predictor gain value is determined by said parameter estimator calculating the following:

$$(y_n^T \cdot x_{f,n-1}) / (x_{f,n-1}^T \cdot x_{f,n-1})$$

wherein, y_n^T is the transpose of said current input vector y_n that is an nth one of said sequence of input vectors; and $x_{f,n-1}^T$ is the transpose of the previous filtered estimate value $x_{f,n-1}$ resulting from a previous input vector y_{n-1} .

Preferably, said filter has a predictor error variance output and an observation noise variance input, said

predictor error variance output providing a current predictor error variance value.

Suitably, said parameter estimator has an observation
 5 noise variance output coupled to said observation noise variance input, and a predictor error variance input coupled to said predictor error variance output, said predictor error variance output providing a current predictor error variance value,

10 wherein when said vector estimation system receives said current input vector, said parameter estimator provides a current observation noise variance value at said observation noise variance output thereby modifying said current filtered estimate value at said current slowly
 15 evolving filter estimate output, said current observation noise variance value being dependent upon said previous filtered estimate value received at said previous slowly evolving filter estimate input, said current input vector received at said estimator vector input, said current
 20 predictor gain value and said current predictor error variance value.

Suitably, said current observation noise variance value is determined by calculating the following:
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$$(y_n^T \cdot (y_n - \alpha_n \cdot x_{f,n-1}) / N) - \Sigma_{p,n}$$

wherein N is a number of elements of said current input vector y_n ; $\Sigma_{p,n}$ is the current predictor error variance value associated with said current input vector y_n ; α_n is
 30 said current predictor gain value; and $x_{f,n-1}$ is said previous filtered estimate value.

Preferably, said parameter estimator has an OnsetFlag output coupled to an OnsetFlag input of said digital

filter, wherein if a signal at said OnsetFlag input is above a threshold value dependent upon harmonic energy in said current input vector, said previous filtered estimate value is set to a filtered estimate value $x_{f,n-1}$.

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Suitably, if said signal at said OnsetFlag input is below the threshold value dependent upon harmonic energy in said current input vector, said previous filtered estimate value is set to a previous input vector y_{n-1} .

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Preferably, the parameter estimator has an unvoiced speech module that determines the current input vector's harmonic energy content by assessing the current predictor gain value and depending upon the current predictor gain value the parameter estimator selectively sets the current predictor gain value and the current observation noise variance value.

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According to another aspect of the invention there is provided a vector estimation system for processing a sequence of input vectors, said input vectors each comprising a plurality of elements, and said system comprising:

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a digital filter with a filter vector input for receiving said sequence of input vectors and an observation noise variance input for controlling characteristics of said filter, said digital filter also having a current slowly evolving filter estimate output, a predictor error variance output and a previous slowly evolving filter estimate output, said current slowly evolving filter estimate output providing a current filtered estimate value of a slowly evolving component of said sequence of input vectors, said predictor error variance output

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providing a current predictor error variance value and said previous slowly evolving filter estimate output providing a previous filtered estimate value of said slowly evolving component of said sequence of input vectors; and

a parameter estimator having an estimator vector input for receiving said sequence of input vectors and a previous slowly evolving filter estimate input coupled to said previous slowly evolving filter estimate output, said parameter estimator further includes a observation noise variance output coupled to said observation noise variance input,

wherein when said vector estimation system receives a current input vector that is one of said sequence of said input vectors, said parameter estimator provides a current observation noise variance value at said observation noise variance output thereby modifying said current filtered estimate value at said current slowly evolving filter estimate output, said current observation noise variance value being dependent upon said current input vector, said current predictor error variance value, and said previous filtered estimate value.

Suitably, said current observation noise variance value is determined by calculating the following:

$$(Y_n^T \cdot (Y_n - x_{f,n-1})/N) - \Sigma_{p,n}$$

wherein Y_n^T is the transpose of said current input vector Y_n that is an nth one of said sequence of input vectors; N is a number of elements of said current input vector Y_n ; $\Sigma_{p,n}$ is the current predictor error variance value associated with said current input vector Y_n ; and $x_{f,n-1}$ is said previous filtered estimate value.

Preferably, said parameter estimator has an OnsetFlag output coupled to an OnsetFlag input of said digital filter, wherein if a signal at said OnsetFlag input is
5 above a threshold value dependent upon harmonic energy in said current input vector, said previous filtered estimate value is set to a filtered estimate value $x_{f,n-1}$.

Suitably, if said signal at said OnsetFlag input is
10 below the threshold value dependent upon harmonic energy in said current input vector, said previous filtered estimate value is set to a previous input vector y_{n-1} .

Preferably, the parameter estimator has an unvoiced
15 speech module that determines the current input vector's harmonic energy content by assessing the current predictor gain value and depending upon the current predictor gain value the parameter estimator selectively sets the current predictor gain value and the current observation noise
20 variance value.

Suitably, said digital filter further includes:
a slowly evolving predicted estimate output providing a
current predicted estimate value of said sequence of
25 vectors. The digital filter may also have a process noise variance input.

Suitably, there may be a smoother module having inputs
coupled respectively to at least two outputs of said
30 digital filter.

Preferably, said smoother module has five inputs coupled to respective outputs of said filter. Preferably, said smoother module has a smoothed estimate output

providing a smoothed estimate value of a previous slowly evolving component.

Suitably, said smoothed estimate output is coupled to
5 a smoothed estimate input of said parameter estimator.

According to another aspect of the invention there is provided a method for processing a sequence of input vectors each comprising a plurality of elements, said
10 vectors being applied to a vector estimation system having a parameter estimator coupled to a digital filter, said method comprising the steps of:

receiving said sequence of input vectors at inputs of said filter and said parameter estimator;
15 determining a current predictor gain value from a current input vector that is one of said sequence of said input vectors, said determining being effected by said parameter estimator;and

applying said current predictor gain value to said digital filter to thereby modify a current filtered estimate value provided at an output of said digital filter, said current predictor gain value being dependent upon a previous filtered estimate value from said filter and said current input vector.
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Preferably, said step of determining may be further characterised by providing a current observation noise variance value and a current predictor error variance value from said current input vector,
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Suitably, said step of applying may be further characterised by said filter receiving current observation noise variance value thereby modifying said current filtered estimate value, said current observation noise
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variance value being dependent upon said previous filtered estimate value, said current input vector, said current predictor gain value and said current predictor error variance value.

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According to another aspect of the invention there is provided a method for processing a sequence of input vectors each comprising a plurality of elements, said vectors being applied to a vector estimation system having a parameter estimator coupled to a digital filter, said method comprising the steps of:

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receiving said sequence of input vectors at inputs of said filter and said parameter estimator;

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determining a current observation noise variance value from a current input vector that is one of said sequence of said input vectors, said determining being effected by said parameter estimator; and

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applying said current observation noise variance value to said digital filter to thereby modify a current filtered estimate value provided at an output of said digital filter, said current observation noise variance value being dependent upon said current input vector, a current predictor error variance value from said filter, and a previous filtered estimate value from said filter.

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Preferably, the filter is a Kalman filter.

According to another aspect of the invention there is provided an encoder for processing a speech signal, said encoder comprising:

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A signal normalization module for processing the speech signal to provide a sequence of input vectors each comprising a plurality of elements;

5 a digital filter with a filter vector input coupled to an output of the signal normalization module for receiving said sequence of input vectors, the digital filter also having an observation noise variance input for controlling characteristics of said filter, said digital filter also having a current slowly evolving filter estimate output, a predictor error variance output and a previous slowly evolving filter estimate output, said current slowly evolving filter estimate output providing a current filtered estimate value of a slowly evolving component of said sequence of input vectors, said predictor error variance output providing a current predictor error variance value and said previous slowly evolving filter estimate output providing a previous filtered estimate value of said slowly evolving component of said sequence of input vectors;

10 a parameter estimator having an estimator vector input for receiving said sequence of input vectors and a previous slowly evolving filter estimate input coupled to said previous slowly evolving filter estimate output, said parameter estimator further includes a observation noise variance output coupled to said observation noise variance input; and

25 a slowly evolving component encoder with an input coupled to said slowly evolving filter estimate output, wherein when said vector estimation system receives a current input vector that is one of said sequence of said input vectors, said parameter estimator provides a current observation noise variance value at said observation noise variance output thereby modifying said current filtered estimate value at said current slowly evolving filter estimate output, said current observation noise

variance value being dependent upon said current input vector, said current predictor error variance value, and said previous filtered estimate value.

- 5 Preferably, the encoder may include an adder module with one input coupled said slowly evolving filter estimate output and another input coupled to the output of the signal normalization module, wherein in use said adder subtracts the said current filtered estimate value at the
- 10 output of the vector estimation system from at least one of the elements of the sequence of input vectors.

Suitably, an output of the adder module is coupled to a rapidly evolving component encoder.

- 15 Suitably, said parameter estimator is characterised by said current observation noise variance value being dependent upon both a sequence of previous said input vectors and a sequence of filtered estimate values provided
- 20 by successive preceding values of said previous filtered estimate value.

BRIEF DESCRIPTION OF THE DRAWINGS

- 25 In order that the invention may be readily understood and put into practical effect, reference will now be made to a preferred embodiment as illustrated with reference to the accompanying drawings in which:
- 30 FIG. 1 illustrates a vector estimation system for processing a sequence of input vectors in accordance with a preferred embodiment of the invention;

FIG. 2 illustrates a digital filter forming part of the vector estimation system of FIG. 1;

5 FIG. 3 illustrates a parameter estimator forming part of the vector estimation system of FIG. 1;

FIG. 4 illustrates a smoother module forming part of the vector estimation system of FIG. 1; and

10 FIG. 5 illustrates a speech encoder that includes the vector estimation system of FIG. 1.

DETAILED DESCRIPTION OF THE PREFERRED EMBODIMENT OF THE INVENTION

15 In the drawings, like numerals on different Figs are used to indicate like elements throughout. Referring to Fig. 1, there is illustrated a vector estimation system 1 for processing a sequence of input vectors (y_0 to y_T). The vector estimation system 1 includes a digital filter 2, a
20 parameter estimator 10 and a smoother module 17. The digital filter 2 has five inputs and five outputs. The five inputs of the digital filter 2 are a filter vector input 3, a predictor gain input 4, an observation noise
25 variance input 5, an OnsetFlag input 26 and process noise variance input 25. The five outputs of the digital filter 2 are a current slowly evolving filter estimate output 6, a current slowly evolving predicted estimate output 7, a
30 previous filter error variance output 9, a previous slowly evolving filter estimate output 20 and a current predictor error variance output 21.

The parameter estimator 10 has four inputs and three outputs. The parameter estimator 10 inputs are an

estimator vector input 19 coupled to the vector input 3; a previous slowly evolving filter estimate input 13 coupled to the previous slowly evolving filter estimate output 20; a current predictor error variance input 15 coupled to the
 5 current predictor error variance output 21; and a smoothed estimate input 16. The three outputs of the parameter estimator 10 are a predictor gain output 11 coupled to the predictor gain input 4; an observation noise variance output 12 coupled to the observation noise variance input
 10 5; and an OnsetFlag output 22 coupled to the OnsetFlag input 26.

The smoother module 17 has six inputs one being coupled to the slowly evolving filter estimate output 6;
 15 one coupled to the slowly evolving predicted estimate output 7; one coupled to the previous filter error variance output 9; one coupled to the previous slowly evolving filter estimate output 20; one coupled to the predictor error variance output 21; and one coupled to the predictor gain output 11. The smoother module 17 also has a smoothed
 20 estimate output 18 providing an output for the vector estimation system 1, the smoothed estimate output 18 is coupled to the smoothed estimate input 16 of the parameter estimator 10.

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Referring to Fig. 2, the digital filter 2 is a comb filter in the form of a Kalman Filter Bank. The digital filter 2 comprises a Kalman gain determination module 30 with one input being the observation noise variance input
 30 5. There is also a filtered state estimation module 31 with one input being the vector input 3 and another input being coupled to an output of the Kalman gain determination module 30. An output from the filtered state estimation module 31 provides the slowly evolving filter estimate

output 6 that is coupled to an input of a previous filtered state adjustment module 33. Other inputs of the previous filtered state adjustment module 33 are provided by the OnsetFlag input 26 and the vector input 3 via a delay module 34.

An output from the previous filtered state adjustment module 33 provides the previous slowly evolving filter estimate output 20 that is coupled to an input of a predicted state estimation module 35. Another input to the predicted state estimation module 35 is provided by the predictor gain input 4. An output of the predicted state estimation module 35 provides the slowly evolving predicted estimate output 7 that is coupled to an input of the filtered state estimation module 31.

The output from the Kalman gain determination module 30 is also coupled to an input of a filter variance estimation module 32 that has an output coupled to an input to a previous filter variance adjustment module 36. An output from the previous filter variance adjustment module 36 provides the previous filter error variance output 9 that also provides an input to a predictor variance estimation module 37. Other inputs to the predictor variance estimation module 37 are provided by the predictor gain input 4, process noise variance input 25, OnsetFlag input 26 and observation noise variance input 5. An output from the predictor variance estimation module 37 provides the predictor error variance output 21 that is coupled to inputs of the Kalman gain determination module 30, the filter variance estimation module 32 and previous filter variance adjustment module 36. Other inputs to the previous filter variance adjustment module 36 are provided

by the predictor gain input 4, the process noise variance input 25 and the OnsetFlag input 26.

- As will be apparent to a person skilled in the art, the characteristics of the digital filter 2 are formalised in equations (1)-(6) below.

- At an n th input vector y_n (a current input vector) of the series of input vectors (y_0 to y_T) received by the system 1, the previous filtered state adjustment module 33 provides, at the previous slowly evolving filter estimate output 20, a previous filtered estimate value $x_{f,n-1}$.

- The *OnsetFlag* input 26 is a binary signal input that indicates whether or not the beginning of a signal segment containing a significant amount of harmonic energy (determined by a threshold value) has been detected. If *OnsetFlag* input 26 is set to a value that indicates that the beginning of such a segment has been detected, then the previous filtered estimate value $x_{f,n-1}$ is set to an a previous input vector y_{n-1} .

- Alternatively, if the *OnsetFlag* input 6 is set to a value that indicates that the beginning of a segment containing a significant amount of harmonic energy has been detected, then the previous filtered estimate value $x_{f,n-1}$ is set to a value of the previous input vector y_{n-1} that was provided at vector input 3.

- For the current input vector y_n , the digital filter 2 provides a current predicted estimate value $x_{p,n}$ at the predicted estimate output 7. The current predicted estimate value $x_{p,n}$ is computed according to:

$$x_{p,n} = \alpha_n \cdot x_{f,n-1} \quad \text{----(1)}$$

Where α_n is a current predictor gain value for the nth input vector y_n , provided at the predictor gain input 4.

5 Once the current predicted estimate value $x_{p,n}$ is computed, then also for the current input vector y_n a current filtered estimate value $x_{f,n}$ is provided at the slowly evolving filter estimate output 6. This current filtered estimate value $x_{f,n}$ is computed according to:

$$10 \quad x_{f,n} = x_{p,n} + K_n \cdot (y_n - x_{p,n}) \quad \text{----(2)}$$

Where K_n is a current Kalman gain value for the digital filter 2 for the nth current input vector y_n .

15 The Kalman gain value K_n is computed according to:

$$K_n = \Sigma_{p,n} / (\Sigma_{p,n} + \sigma_{v,n}^2) \quad \text{----(3)}$$

20 Where, $\Sigma_{p,n}$ is a current predictor error variance value provided at the predictor error variance output 21 for the nth input vector y_n ; and $\sigma_{v,n}^2$ is a current observation noise variance value provided at the observation noise variance input 5 also for the nth input
25 vector y_n .

If the *OnsetFlag* is set to a value that indicates that the beginning of a signal segment containing a significant amount of harmonic energy has been detected, then the
30 current predictor error variance value $\Sigma_{p,n}$ is typically set to the observation noise variance value $\sigma_{v,n}^2$. This results in Equation (3) producing a Kalman gain value K_n equal to 0.5.

- If *OnsetFlag* is set to a value that indicates that the beginning of a signal segment containing a significant amount of harmonic energy has not been detected, then the
- 5 current predictor error variance value $\Sigma_{p,n}$ is computed according to:

$$\Sigma_{p,n} = \alpha_n \cdot \alpha_n \cdot \Sigma_{f,n-1} + \sigma_w^2 \quad \text{----(4)}$$

- 10 where σ_w^2 is a process noise variance value provided at the process noise variance input 25 and $\Sigma_{f,n-1}$ is a previous filtered error variance value for a previous input vector y_{n-1} .

- 15 If the *OnsetFlag* is set to a value that indicates that the beginning of a signal segment containing a significant amount of harmonic energy has not been detected then a current filtered error variance value $\Sigma_{f,n}$, provided at the output of the filter error variance estimation module 32,
- 20 is computed according to:

$$\Sigma_{f,n} = (1 - K_n) \cdot \Sigma_{p,n} \quad \text{----(5)}$$

- If the *OnsetFlag* is set to a value that indicates that
- 25 the beginning of a signal segment containing a significant amount of harmonic energy has been detected, then the current filtered error variance value $\Sigma_{f,n}$ is computed according to:

- 30
$$\Sigma_{f,n} = (\Sigma_{p,n} - \sigma_w^2) / \alpha_n^2 \quad \text{----(6)}$$

Referring to Fig. 3 there is illustrated the parameter estimator 10 that typically comprises an initial parameter

estimation module 40, an unvoiced speech adjustment module 41 and a voicing onset adjustment module 42. The initial parameter estimation module 40 has four inputs provided by the predictor error variance input 15, the slowly evolving filter estimate input 13, the vector input 19 and the smoothed estimate input 16. Outputs of the parameter estimation module 40 are coupled to inputs of the unvoiced speech adjustment module 41 and a further inputs to module 41 are provided by the predictor error variance input 15, the slowly evolving filter estimate input 13 and the vector input 19. Outputs of the unvoiced speech adjustment module 41 are coupled to inputs the voicing onset adjustment module 42 and a further input to module 42 is provided by the vector input 3. The voicing onset adjustment module 42 has three outputs providing the predictor gain output 11, observation noise variance output 12 and OnsetFlag output 22.

The initial parameter estimation module 40 computes initial estimates of the current predictor gain value α_{2^n} and the current observation noise variance value σ_{vn} . These are determined as follows:

$$\alpha_n = (y_n^T \cdot x_{f,n-1}) / (x_{f,n-1}^T \cdot x_{f,n-1}) \quad \text{--- (7)}$$

$$\sigma_{vn}^2 = (y_n^T \cdot (y_n - \alpha_n \cdot x_{f,n-1}) / N) - \Sigma p_n \quad \text{--- (8)}$$

where N is the number of elements in the current input vector y_n that is the nth one of the sequence of input vectors $(y_0$ to $y_T)$ received at vector input 3; y_n^T is the transpose of the current input vector y_n ; and $x_{f,n-1}^T$ is the transpose of the previous filtered estimate value $x_{f,n-1}$ resulting from the previous input vector y_{n-1} applied to the digital filter system 1. These estimate values are

based on the assumption that both the current and at least the previous two input vectors represent signal segments containing a significant amount of harmonic energy.

- 5 It should be noted that for certain applications, the current predictor gain value α_n may be set to a constant value of 1. Thus, current observation noise variance value σ_{vn}^2 is determined as follows:

$$10 \quad \sigma_{vn}^2 = (y_n^T \cdot (y_n - x_{f,n-1}) / N) - \Sigma_{p,n} \quad \text{-----(8a)}$$

- The unvoiced speech adjustment module 41 determines whether the current input vector y_n represents a segment of
 15 speech that contains no significant harmonic energy, and if so selectively sets the current predictor gain value α_n and the current observation noise variance value σ_{vn}^2 to appropriate values. Preferably, the unvoiced speech adjustment unit determines that the current input vector y_n
 20 represents a segment of speech that contains no significant harmonic energy by detecting whether either of the following conditions is true:

- (i) α_n is less than 0.0, or
 25 (ii) both α_n is greater than 1.0 and the initial estimation of the observation noise variance value is greater than mean squared value of elements in the current predicted estimate value.

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If either conditions (i) or (ii) hold, then typically the unvoiced speech adjustment module 41 will set α_n to 1.0, and re-compute σ_{vn}^2 accordingly using Equation (8).

The voicing onset adjustment module 42 determines if the current input vector y_n represents the second cycle of a segment of speech containing a significant amount of harmonic energy, and if so adjusts current predictor gain value α_n and the observation noise variance value σ_{vn}^2 to more appropriate values and sets the *OnsetFlag* to a value indicating that voicing onset has been detected.

Typically, the voicing onset adjustment module 42 determines that the current input vector y_n is the second cycle of a segment of speech containing a significant amount of harmonic energy as follows. An input prediction gain, β , is computed according to:

$$\beta = (y_n^T \cdot y_{n-1}) / (y_{n-1}^T \cdot y_{n-1}) \quad \text{---(9)}$$

An input prediction error variance value, σ_e^2 , is computed according to:

$$\sigma_e^2 = y_n^T \cdot (y_n - \beta \cdot y_{n-1}) / N \quad \text{---(10)}$$

The voicing onset adjustment unit determines whether both of the following conditions are true:

- (iii) σ_e^2 is less than $k_1 \cdot \sigma_{vn}^2$, wherein k_1 is a constant, whose value is typically 0.9.
- (iv) σ_e^2 divided by the mean squared value of the elements of the input vector is less than k_2 , wherein k_2 is a constant, whose value is typically 0.5.

If both conditions (iii) and (iv) hold, then typically the voicing onset adjustment unit will set α_n to β and set σ_{vn}^2 to σ_e^2 .

Referring to Fig. 4 there is illustrated the smoother module 17 that typically comprises series coupled smoothed

state estimation modules 50 a first stage of which has an input receiving the current filtered estimate value $x_{f,n}$. The final stage of the smoothed state estimation module 50 provides a smoothed estimate value $x_{f,n-j}$ at output 18 of a previous slowly evolving component. The smoother module 17 also has five sets of series coupled delay modules 51,52,53,54 and 55 with respective outputs of an j th delay module 51,52,53,54 and 55 providing inputs to an j +1th smoothed state estimation module 50.

The smoothed state estimation modules 50 provide smoothed estimates $X_{s,(n-j)}$ for successive values of j beginning with $j=1$. These estimates are computed according to:

$$X_{s,(n-j)} = x_{f,(n-j)} + C \cdot (x_{s,(n-j+1)} - X_{p,(n-j+1)}) \quad \text{----(11)}$$

wherein

$$C = (\Sigma_{f,-j} \cdot \alpha_{n-j} / \Sigma_{p,(n-j+1)}) \quad \text{----(12)}$$

20 and

$$X_{s,n} = X_{f,n} \quad \text{----(13)}$$

From the above it will be apparent that the purpose of the smoother module 17 is to provide an estimate $X_{s,(n-j)}$ of the slowly evolving component of an input vector y_{n-j} based upon input vectors up to and including y_n . The smoother module 17 thus uses current data to estimate a past slowly evolving component value, in contrast to the digital filter 2, which uses current data to estimate a current slowly evolving component value.

In use, the vector estimation system 1 receives the sequence of input vectors y_0 to y_T that are each comprising N elements. Each of the input vectors y_0 to y_T contains a sampled period of a presumed quasiperiodic signal. This
 5 sampled signal is typically time warped to allow for variations of quasiperiodic periods, so that each input vector contains the same number of elements, as will be apparent to a person skilled in the art. Alternatively, consecutive input vectors y_0 to y_T may have elements added
 10 to them or removed from them, again so that the resulting number of elements in each is the same. For an n th iteration, a input vector y_n will be applied to vector input 3 and estimator vector input 19. The digital filter 2 processes this input vector y_n resulting in the slowly
 15 evolving filter estimate output 6 providing, to input 13, the previous filtered estimate value $x_{f,n-1}$ of a slowly evolving component of sequence of vectors y_0 to y_T .

The parameter estimator 10 processes the previous
 20 filtered estimate value $x_{f,n-1}$ and current input vector y_n to provide a current current predictor gain value α_n at predictor output 11. The current predictor gain value α_n is thereby applied to input 4 of the digital filter 2 for controlling the gain thereof during filtering of input
 25 vector y_n . The parameter estimator 10 determines the current predictor gain value α_n by the calculation stated in equation (7).

As will be apparent to a person skilled in the art, at
 30 initialisation (i.e. the first sample time when n is 0 therefore input vector y_0 is applied to digital filter system 1), there will be no previous filtered estimate value $x_{f,n-1}$. Accordingly, although there are many ways to allocate a value for this previous filtered estimate value

$x_{f,n-1}$, the present invention preferably assigns the previous filtered estimate value $x_{f,n-1}$ with the same element values as input vector y_0 .

5 Referring to Fig. 5, the vector estimation system 1 can advantageously included in a speech encoder 60. The speech encoder 60 includes a signal normalization module 61 with an input for receiving a speech signal. A signal vector output 62 of the signal normalization module 61 is coupled to filter vector input 3 and another output is coupled to the process noise variance input 25 of the vector estimation system 1. The signal vector output 62 is also coupled to an input of an adder module 63 and another input of the adder module 63 is coupled to the smoothed estimate output 18 of the vector estimation system 1. An output from the adder module 63 is coupled to an input of a rapidly evolving component encoder 64 and there is also a slowly evolving component encoder 65 having an input coupled to the smoothed estimate output 18. The speech encoder 60 has three outputs from the signal normalization module 61, for coupling to a speech decoder, these outputs being a spectral envelope output 66, an energy parameters output 67 and a period parameters output 68. The speech encoder 60 also has a slowly evolving component output 69 from the evolving component encoder 65 and a rapidly evolving component output 70 from the rapidly evolving component encoder 64.

30 In operation, the speech encoder 60 firstly normalizes a speech signal with respect to its spectral envelope, energy and period. The normalisation process involves estimating parameters that describe the spectral envelope, energy and period of the input signal and these parameters are typically transmitted to a speech decoder at outputs

66,67,68. Process noise variance provide at the process noise variance input 25 is typically used to control the vector estimation system 1. The normalisation process produces the sequence of input vectors (y_0 to y_T) for the vector estimation system 1. The sequence of input vectors (y_0 to y_T) are a sequence of fixed length vectors representing sampled consecutive cycles of the normalised waveform. These vectors (y_0 to y_T) are applied to the filter vector input 3 of the vector estimation system 1, which generates a slowly evolving component at the smoothed estimate output 18. By subtracting this slowly evolving component from the input vectors (y_0 to y_T), a rapidly evolving, or noise-like component is produced and provided to the rapidly evolving component encoder 64. The slowly evolving and rapidly evolving components are encoded respectively by the slowly and rapidly evolving component encoders 64, 65. The encoders 64,65 use appropriate methods known in the art to produce parameters at respective outputs 70,69 which are transmitted to a speech decoder.

Advantageously, as can be seen with reference to equation (7), the present invention provides for the vector estimation system 1 to receive the current input vector y_n that is one of a the sequence of input vectors y_0 to y_T . The parameter estimator 10 provides the current predictor gain value α_n at the predictor gain output 11 thereby modifying the current filtered estimate value $x_{f,n}$ at the slowly evolving filter estimate output 6. In this regard, the current predictor gain value α_n is dependent upon the previous filtered estimate value $x_{f,n-1}$ and the current input vector y_n . As will be apparent to a person skilled in the art, the parameter estimator 10 determines the current predictor gain value α_n from both a sequence of

previous input vectors y_{n-1} to y_0 and a sequence of corresponding filtered estimate values provided by successive preceding values of the previous filtered estimate value $x_{f,0}$ to $x_{f,n-1}$.

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With reference to equation (8), the present invention also advantageously provides for the vector estimation system 1 to receive the current input vector y_n . The parameter estimator 10 provides the current observation noise variance value $\sigma_{v,n}^2$ at the observation noise variance output 12 thereby modifying current filtered estimate value $x_{f,n}$ at the slowly evolving filter estimate output 6. The current observation noise variance value $\sigma_{v,n}^2$ is dependent upon the current input vector y_n , the current predictor gain value α_n (that can be set to a constant value of 1 for certain applications), the current predictor error variance value $\Sigma_{p,n}$, and said previous filtered estimate value $x_{f,n-1}$.

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The detailed description provides a preferred exemplary embodiment only, and is not intended to limit the scope, applicability, or configuration of the invention. Rather, the detailed description of the preferred exemplary embodiment provides those skilled in the art with an enabling description for implementing a preferred exemplary embodiment of the invention. It should be understood that various changes may be made in the function and arrangement of elements without departing from the spirit and scope of the invention as set forth in the appended claims.

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